



# POSTAL BOOK PACKAGE 2024

## ELECTRONICS ENGINEERING

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### CONVENTIONAL Practice Sets

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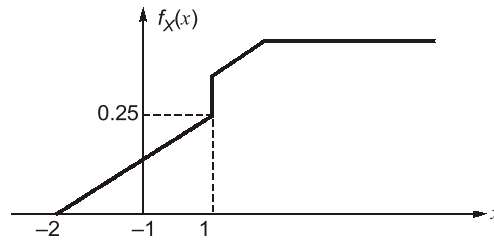
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# Theory of Random Variable and Noise

**Q1** Define PDF and summarise its important properties. Also calculate the probability of outcome of a Random Variable (RV)  $X$  having  $X \leq 1$  for the following PDF curve of RV as shown.



**Solution:**

Probability density function specifies the probability of a random variable taking a particular value.

The Probability Density Function (PDF) which is generally denoted by  $f_X(x)$  or  $P_X(x)$  or  $p_X(x)$  is defined in terms of the Cumulative Distribution Function (CDF)  $F_X(x)$  as,

$$\text{PDF} = f_X(x) = \frac{d}{dx} F_X(x) \quad \dots(i)$$

**The PDF has the following properties:**

- (i)  $f_X(x) \geq 0$  for all  $x$

This results from the fact that probability cannot be negative. Also,  $F_X(x)$  increases monotonically, as  $x$  increases, more outcomes are included in the prob. of occurrence represented by  $F_X(x)$ .

- (ii) Area under the PDF curve is always equal to unity.

i.e. 
$$\int_{-\infty}^{\infty} f_X(x) dx = 1$$

- (iii) The CDF is obtained by the result

$$\text{CDF} = \int_{-\infty}^x f_X(x) dx$$

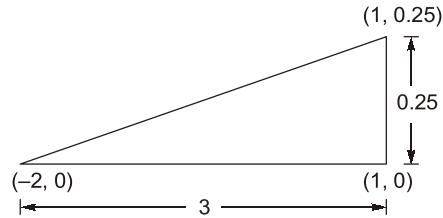
- (iv) Probability of occurrence of the value of random variable between the limits of  $x_1$  and  $x_2$  is given by,

$$P(x_1 < X < x_2) = \int_{x_1}^{x_2} f_X(x) dx$$

Now consider the given PDF curve, since we have to find  $P(x \leq 1)$  so,

Equation for the PDF curve for  $x \leq 1$  is,

$$f_X(x) = \left( \frac{1}{12}x + \frac{1}{6} \right)$$



Now,  $P(x \leq 1)$

$$= P(-2 < x < 1) = \int_{-2}^1 \left( \frac{1}{12}x + \frac{1}{6} \right) dx = \left[ \frac{1}{12} \cdot \frac{x^2}{2} + \frac{1}{6}x \right]_{-2}^1 = \frac{3}{8}$$

$$\therefore P(x \leq 1) = \frac{3}{8}$$

**Q2** Find the cumulative distribution function  $F(x)$  corresponding to the PDF  $f(x) = \frac{1}{\pi(1+x^2)}$ ,  $-\infty < x < \infty$ .

**Solution:**

Given  $f(x) = \frac{1}{\pi(1+x^2)}$ ,  $-\infty < x < \infty$

$$F(x) = P(X \leq x)$$

$$= \int_{-\infty}^x f(x) dx = \frac{1}{\pi} \int_{-\infty}^x \frac{dx}{1+x^2} = \frac{1}{\pi} [\tan^{-1} x]_{-\infty}^x = \frac{1}{\pi} \left( \frac{\pi}{2} + \tan^{-1} x \right)$$

**Q3** Given the random variable  $X$  with density function

$$f_X(x) = \begin{cases} 2x, & 0 < x < 1 \\ 0, & \text{otherwise} \end{cases}$$

Find the PDF of  $Y = 8X^3$ .

**Solution:**

$y = 8x^3$  is an increasing function in  $(0, 1)$

Given  $y = 8x^3$

$$\Rightarrow x^3 = \frac{y}{8}$$

$$\Rightarrow x = \left( \frac{y}{8} \right)^{1/3} = \frac{1}{2} y^{1/3}$$

and  $f_X(x) = 2x$ ,  $0 < x < 1$

$$f_X(y) = \frac{2y^{1/3}}{2} = \frac{y^{1/3}}{3}$$

$$f_Y(y) =$$

...(i)

Given  $x = \left( \frac{y}{8} \right)^{1/3} = \frac{1}{2} y^{1/3} \Rightarrow \frac{dx}{dy} = \frac{1}{6} y^{-2/3}$

Using it in (i)

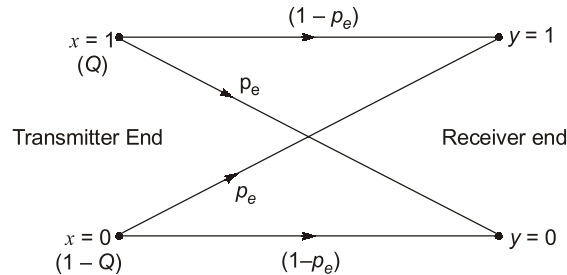
$$f_Y(y) = y^{1/3} \cdot \frac{1}{6} y^{-2/3} = \frac{1}{6} y^{-1/3} = \frac{1}{6} \frac{1}{y^{1/3}} = \frac{1}{6 \sqrt[3]{y}}$$

The range for  $x$  is  $0 < x < 1$

When  $x = 0$ ,  $y = 8 \times 0 = 0$  and  $x = 1$ ,  $y = 8 \times 1^3 = 8$

$$f_Y(y) = \frac{1}{6\sqrt[3]{y}}, \quad 0 < y < 8$$

**Q4** A BSC (Binary Symmetric Channel) error probability is  $P_e$ . The probability of transmitting '1' is  $Q$ , and that of transmitting '0' is  $(1 - Q)$  as in figure below. Calculate the probabilities of receiving 1 and 0 at the receiver?



**Solution:**

If  $x$  and  $y$  are the transmitted digit and the received digit respectively, then for a BSC,

$$P_{y|x}(0|1) = P_{y|x}(1|0) = P_e$$

$$P_{y|x}(0|0) = P_{y|x}(1|1) = 1 - P_e$$

Also,

$$P_x(1) = Q \text{ and } P_x(0) = 1 - Q$$

We have to find,  $P_y(1)$  and  $P_y(0)$  = ?

$$\therefore P_y(1) = P_x(0) P_{y|x}(1|0) + P_x(1) P_{y|x}(1|1) = (1 - Q)P_e + Q(1 - P_e)$$

$$\text{also, } P_y(0) = P_x(0) P_{y|x}(0|0) + P_x(1) P_{y|x}(0|1) = (1 - Q)(1 - P_e) + QP_e$$

**Q5** For the triangular distribution

$$f(x) = \begin{cases} x, & 0 \leq x \leq 1 \\ 2 - x, & 1 \leq x \leq 2 \\ 0, & \text{otherwise} \end{cases}$$

Find the mean and variance.

**Solution:**

$$\text{Mean} = E(X) = \int_{-\infty}^{\infty} xf(x)dx = \int_0^1 x \cdot x dx + \int_1^2 x(2-x)dx = \int_0^1 x^2 dx + \int_1^2 (2x - x^2)dx$$

$$= \left[ \frac{x^3}{3} \right]_0^1 + \left[ 2\left(\frac{x^2}{2}\right) - \frac{x^3}{3} \right]_1^2$$

$$= \frac{1}{3} + \left[ \left(4 - \frac{8}{3}\right) - \left(1 - \frac{1}{3}\right) \right] = \frac{1}{3} + \frac{4}{3} - \frac{2}{3} = 1$$

$$E(X^2) = \int_{-\infty}^{\infty} x^2 f(x)dx = \int_0^1 x^2 \cdot x dx + \int_1^2 x^2 (2-x)dx$$

$$= \int_0^1 x^3 dx + \int_1^2 (2x^2 - x^3)dx = \left[ \frac{x^4}{4} \right]_0^1 + \left[ 2\left(\frac{x^3}{3}\right) - \frac{x^4}{4} \right]_1^2$$

$$= \frac{1}{4} + \left[ \left(\frac{16}{3} - \frac{16}{4}\right) - \left(\frac{2}{3} - \frac{1}{4}\right) \right] = \frac{1}{4} + \frac{16}{3} - 4 - \frac{2}{3} + \frac{1}{4} = \frac{7}{6}$$

$$\text{Var}(X) = E(X^2) - E(X)^2 = \frac{7}{6} - (1)^2 = \frac{1}{6}$$

**Q6** The joint density function of two continuous random variables is given by

$$f(x, y) = \begin{cases} xy/8, & 0 < x < 2, 1 < y < 3 \\ 0, & \text{otherwise} \end{cases}$$

Find (a)  $E(X)$ , (b)  $E(Y)$  and (c)  $E(2X + 2Y)$ .

**Solution:**

$$(a) \quad E(X) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xf(x, y) dx dy = \int_{x=0}^2 \int_{y=1}^3 x(xy/8) dx dy = \frac{4}{3}$$

$$(b) \quad E(Y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} yf(x, y) dx dy = \int_{x=0}^2 \int_{y=1}^3 y(xy/8) dx dy = \frac{13}{6}$$

$$(c) \quad E(2X + 3Y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (2x + 3y) dx dy = \int_{x=0}^2 \int_{y=1}^3 (2x + 3y)(xy/8) dx dy = \frac{55}{6}$$

**Q7** Let  $z$  be a random variable with probability density function  $f_z(z) = \frac{1}{2}$  in the range  $-1 \leq z \leq 1$ . Let the

random variable  $x = z$  and the random variable  $y = z^2$ . Obviously  $x$  and  $y$  are not independent since  $x^2 = y$ . Show that  $x$  and  $y$  are uncorrelated.

**Solution:**

$$\text{We have,} \quad E(z) = \int_{-1}^1 z \cdot f_z(z) dz$$

$$\Rightarrow \quad E(z) = \frac{1}{4} [z^2]_{-1}^1 = 0$$

$$\text{Since, } x = z, \text{ so } E(x) = E(z) = 0$$

$$\text{Since, } y = z^2 \text{ so } E(y) = E(z^2)$$

$$\text{So that,} \quad E(y) = \int_{-1}^1 \frac{1}{2} z^2 dz = \frac{1}{6} [z^3]_{-1}^1 = \frac{1}{3}$$

We know that, the co-variance ' $\mu$ ' of two RVs  $x$  and  $y$  is defined as,

$$\begin{aligned} \mu &= E\{(x - m_x)(y - m_y)\} \\ &= E\left\{x\left(y - \frac{1}{3}\right)\right\} = E\left\{xy - \frac{1}{3}x\right\} = E\left\{z^3 - \frac{z}{3}\right\} = \int_{-1}^1 \frac{1}{2} \left(z^3 - \frac{z}{3}\right) dz \\ \mu &= 0 \end{aligned}$$

Now, correlation coefficient between the variables  $x$  and  $y$  is defined by quantity ' $\rho$ ' as,

$$\rho = \frac{\mu}{\sigma_x \sigma_y} = 0$$

So, we can say that these RV's  $X$  and  $Y$  are uncorrelated.

**Q8** A WSS random process  $x(t)$  is applied to the input of an LTI system with impulse response

$$h(t) = 3e^{-2t} u(t)$$

Find the mean value of the output  $y(t)$  of the system, if  $E[x(t)] = 2$ . Here  $E[\cdot]$  denotes the expectation operator.

**Solution:**

The output  $y(t)$  is the convolution of the input  $x(t)$  and the impulse response  $h(t)$ .

$$y(t) = \int_{-\infty}^{\infty} h(\tau) \cdot x(t - \tau) \cdot d\tau$$

$$\therefore E[y(t)] = \int_{-\infty}^{\infty} h(\tau) \cdot E[x(t - \tau)] \cdot d\tau$$

$$E[y(t)] = H(0) \times E[x(t)]$$

$$\boxed{E[y(t)] = E[x(t)] \cdot H(0)}$$

where,  $H(0) = H(\omega)|_{\omega=0}$  and  $H(\omega)$  = Fourier transform of  $h(t)$

Given  $E[x(t)] = 2$ ,  $h(t) = 3e^{-2t}u(t)$

Taking Fourier transform,  $H(\omega) = \frac{3}{2 + j\omega} \Rightarrow H(0) = \frac{3}{2}$

$$E[y(t)] = 2 \times \frac{3}{2} = 3$$

**Q9** Suppose that two signals  $s_1(t)$  and  $s_2(t)$  are orthogonal over the interval  $(0, T)$ . A sample function  $n(t)$  of a zero-mean white noise process is correlated with  $s_1(t)$  and  $s_2(t)$  separately, to yield the following variables:

$$n_1 = \int_0^T s_1(t) n(t) dt \quad \text{and} \quad n_2 = \int_0^T s_2(t) n(t) dt$$

Prove that  $n_1$  and  $n_2$  are orthogonal.

**Solution:**

$$\begin{aligned} E[n_1 n_2] &= E \left[ \int_0^T s_1(u) n(u) du \int_0^T s_2(v) n(v) dv \right] \\ &= \int_0^T \int_0^T s_1(u) s_2(v) E[n(u) n(v)] du dv \end{aligned}$$

$n(t)$  is a white noise process.

So,  $R_N(\tau) = \frac{N_0}{2} \delta(\tau)$

$$E[n(u) n(v)] = \frac{N_0}{2} \delta(u - v)$$

Hence,  $E[n_1 n_2] = \frac{N_0}{2} \int_0^T \int_0^T s_1(u) s_2(v) \delta(u - v) du dv$

$$= \frac{N_0}{2} \int_0^T s_1(u) s_2(u) du$$

$$= 0 \quad \because s_1(t) \text{ and } s_2(t) \text{ are orthogonal over } (0, T)$$

$E[n_1 n_2] = 0$ . So,  $n_1$  and  $n_2$  are also orthogonal.

**Q.10** Find the time autocorrelation function of the signal  $g(t) = e^{-at} u(t)$  and from this obtain the energy spectral density (ESD) of  $g(t)$ .

**Solution:**

Auto correlation function,

$$\begin{aligned} R_x(\tau) &= \int_{-\infty}^{\infty} g(t) g(t-\tau) dt = \int_{-\infty}^{\infty} e^{-at} u(t) e^{-a(t-\tau)} u(t-\tau) dt \\ &= \int_{\tau}^{\infty} e^{-2at} e^{a\tau} dt = e^{a\tau} \int_{\tau}^{\infty} e^{-2at} dt = \frac{e^{a\tau}}{-2a} \left[ e^{-2at} \right]_{\tau}^{\infty} = \frac{e^{a\tau}}{2a} e^{-2a\tau} = \frac{e^{-a\tau}}{2a} \end{aligned}$$

Similar process is valid for negative side because for real  $g(t)$ ,  $R_x(\tau)$  is even function of time

$$R_x(\tau) = \frac{e^{-a|\tau|}}{2a}$$

Now we know that

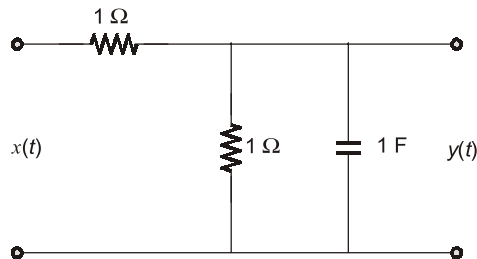
$$R_x(\tau) \xrightarrow{F.T} S_x(\omega)$$

Energy spectral density

$$S_x(\omega) = \frac{1}{\omega^2 + a^2} \triangleq |G(\omega)|^2, \text{ where } G(\omega) = F.T[g(t)]$$

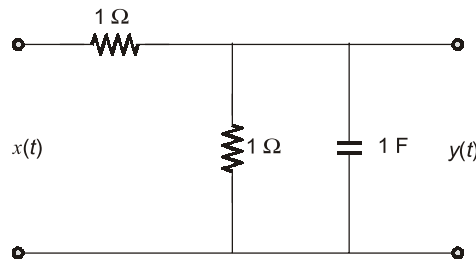
**Q.11** If the input to a low-pass filter as shown below in Figure is a random process  $x(t)$  with autocorrelation function,  $R_x(\tau) = 5\delta(\tau)$ , then find

- Power spectral density of the output random process;
- Average power of output random process.



**Solution:**

Given low pass filter,



By taking Laplace transform

$$\text{Transfer function, } H(s) = \frac{1}{s+2}; \quad H(j\omega) = \frac{1}{j\omega+2}$$

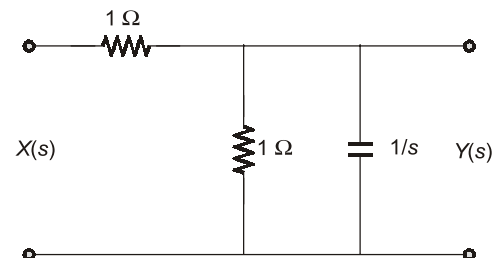
Input Auto-correlation function

$$R_x(\tau) = 5\delta(\tau)$$

Power spectral density  $S_x(\omega) = F[R_x(\tau)] = 5$

Output power spectral density

$$S_y(\omega) = |H(j\omega)|^2 S_x(\omega)$$



$$S_y(\omega) = \frac{5}{\omega^2 + 4}$$

$$\text{Output Auto-correlation function} = R_y(\tau) = \frac{5}{4} e^{-2|\tau|}$$

$$\therefore e^{-|a|t} \xrightarrow{\text{CTFT}} \frac{2a}{\omega^2 + a^2}$$

$$(ii) \text{ Average power of the output process} = R_y(0)$$

$$P_y = \frac{5}{4} \text{ W}$$

**Q.12** It is desired to generate a random signal  $X(t)$  with auto correlation function,  $R_x(\tau) = 5\eta e^{-5|\tau|}$ , by passing a white noise  $n(t)$ , with power spectral density  $S_n(f) = \eta/2$  W/Hz, through an LTI system. Obtain an expression for the transfer function  $H(f)$  of the given LTI system.

**Solution:**

$$\text{Given that, } R_x(\tau) = 5\eta e^{-5|\tau|}$$

So, the spectral density = Fourier transform of  $R_x(\tau)$

$$\begin{aligned} \Rightarrow S_X(\omega) &= \int_{-\infty}^{\infty} R_x(\tau) e^{-j\omega\tau} d\tau = \int_{-\infty}^{\infty} 5\eta e^{-5|\tau|} e^{-j\omega\tau} d\tau \\ &= 5\eta \left[ \int_0^{\infty} e^{-5\tau} e^{-j\omega\tau} d\tau + \int_0^{\infty} e^{-5\tau} e^{j\omega\tau} d\tau \right] \\ &= 5\eta \left[ \frac{e^{-(5+j\omega)\tau}}{-(5+j\omega)} \right]_0^{\infty} + 5\eta \left[ \frac{e^{-(5-j\omega)\tau}}{-(5-j\omega)} \right]_0^{\infty} = 5\eta \left[ \frac{1}{5+j\omega} + \frac{1}{5-j\omega} \right] \end{aligned}$$

$$\therefore S_X(\omega) = \frac{5\eta \times 10}{25 + \omega^2} = \frac{50\eta}{25 + \omega^2} \quad \dots(i)$$

$$\text{Hence, } S_X(\omega) = |H(\omega)|^2 \cdot \frac{\eta}{2}$$

$$\therefore |H(\omega)|^2 = S_X(\omega) \cdot \frac{2}{\eta} = \frac{100\eta}{25 + \omega^2} \cdot \frac{1}{\eta}$$

$$\Rightarrow |H(f)|^2 = \frac{100}{25 + 4\pi^2 f^2}$$

$$\therefore |H(f)| = \frac{10}{\sqrt{25 + 4\pi^2 f^2}} \Rightarrow H(f) = \frac{10}{5 + j2\pi f}$$

**Q.13** The power spectral density of a real stationary random process  $X(t)$  is given by

$$S_X(f) = \begin{cases} \frac{1}{W}, & |f| \leq W \\ 0 & |f| > W \end{cases}$$

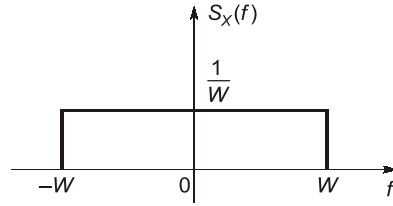
$$\text{Find the value of the expectation } E\left[\pi X(t) \cdot X\left(t - \frac{1}{4W}\right)\right].$$

**Solution:**

$$S_X(f) = \begin{cases} \frac{1}{W}, & |f| \leq W \\ 0 & |f| > W \end{cases}$$



∴



∴

$$R_X(\tau) \xleftrightarrow{\text{CTFT}} [S_X(f)]$$

$$R_X(\tau) = 2\text{sinc}[2W\tau]$$

Also,

$$E\left[\pi X(t) \cdot X\left(t - \frac{1}{4W}\right)\right] = \pi E\left[x(t) \cdot x\left(t - \frac{1}{4W}\right)\right]$$

$$= \pi R_X\left(\frac{1}{4W}\right) = \pi \cdot 2\text{sinc}\left(\frac{1}{2}\right) \quad \dots(i)$$

We know that,

$$\text{sinc}(x) = \frac{\sin \pi x}{\pi x}$$

From equation (i), we get

$$\therefore E\left[\pi X(t) \cdot X\left(t - \frac{1}{4W}\right)\right] = 2\pi \frac{\sin \frac{\pi}{2}}{\frac{\pi}{2}} = 4$$

**Q.14** Stationary random process  $X(t)$  has the following auto correlation function

$$R_X(\tau) = \sigma^2 e^{-\mu|\tau|}$$

where  $\mu$  and  $\sigma^2$  are constants. It is passed through a filter whose impulse response is

$$h(\tau) = \alpha e^{-\alpha\tau} u(\tau)$$

here  $\alpha$  is constant and  $u(\tau)$  is step function.

(a) Find power spectral density of random signal  $X(t)$ .

(b) Find the power spectral density of the output random signal  $Y(t)$ .

**Solution:**

(a) Power spectral density (PSD) of input is expressed as

$$S_X(\omega) = \text{CTFT} \{R_X(\tau)\} = \int_{-\infty}^{\infty} R_X(\tau) e^{-j\omega\tau} d\tau$$

or

$$S_X(\omega) = \int_{-\infty}^{\infty} \sigma^2 e^{-\mu|\tau|} e^{-j\omega\tau} d\tau = \sigma^2 \int_{-\infty}^0 e^{(\mu - j\omega)\tau} d\tau + \sigma^2 \int_0^{\infty} e^{-(\mu + j\omega)\tau} d\tau$$

$$= \sigma^2 \left[ \frac{1}{\mu - j\omega} + \frac{1}{\mu + j\omega} \right] = \frac{2\mu \sigma^2}{\mu^2 + \omega^2}$$

(b) Power spectral density of output is related to power spectral density of input as

$$S_Y(\omega) = |H(\omega)|^2 S_X(\omega) \quad \dots(i)$$

and

$$H(e^{j\omega}) = \text{CTFT} \{h(\tau)\}$$

$$= \int_{-\infty}^{\infty} h(\tau) e^{-j\omega\tau} d\tau = \int_{-\infty}^{\infty} \alpha e^{-\alpha\tau} u(\tau) e^{-j\omega\tau} d\tau$$